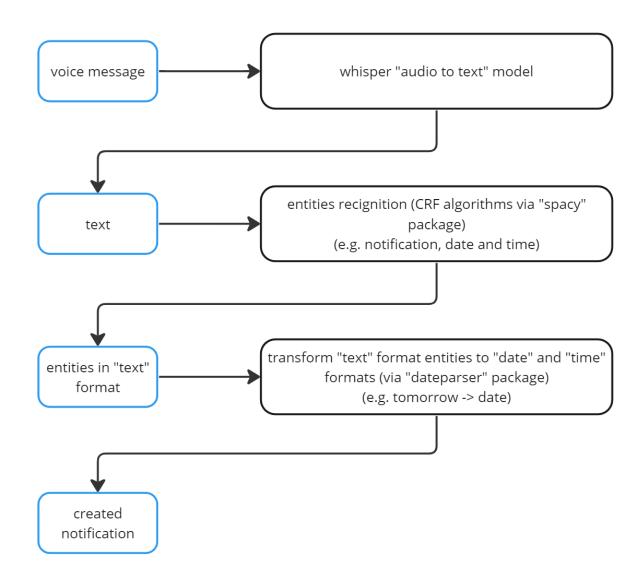
#### **DOCUMENTATION**

## 1) Program logic



#### 2) How spaCy implements NER

spaCy's NER model is based on a machine learning algorithm known as a conditional random field (CRF). The model takes in a sequence of words (tokens) as input and outputs a sequence of labels indicating whether each token is part of a named entity and, if so, which type of entity it belongs to.

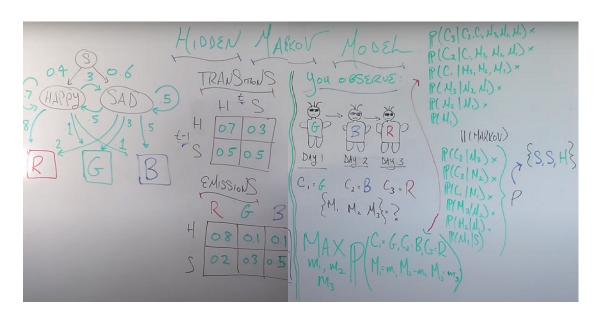
spaCy's NER model is pre-trained on a large corpus of text and can recognize a wide range of named entities out-of-the-box, including people, organizations, locations, and dates. The model also includes additional entity types, such as product names, languages, and nationalities.

#### 3) How CRF works

Actually, CRF is the generalization of the Hidden Markov Model (HMM).

# a) HMM intuitive explanation (Example)

https://www.youtube.com/watch?v=fX5bYmnHqqE&list=LL&index= 34



- hidden entities: moods {happy, sad}. We should guess them by:
- t-shirt color: {R, G, B}. We can see the sequence of them (e.g. during 3 days)
- transitions matrix we get after learning
- emissions matrix we get after learning
- prediction: maximizing the target function:

$$max_{m1,m2,m3}P(C_1 = R, C_2 = G, C_3 = B, M_1 = m_1, M_2 = m_2, M_3 = m_3)$$

- if we make 2 assumptions: colors depend only on layers; current state depends only on previous (Markov condition) => we can compute the value quickly.

### b) HMM mathematical explanation

https://en.wikipedia.org/wiki/Hidden Markov model

# c) CRF implementation

https://www.youtube.com/watch?v=rI3DQS0P2fk&list=LL&index=3
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Actually, everything is the same except assumptions => more complicated computations

## 4) Results analysis

# a) Metrics (for example for "NTFY" entity:

- i) TP: the number of items that marked as "NTFY" in X\_test dataset and exist in y\_test
- ii) FP: the number of items that marked as "NTFY" in X\_test dataset and do NOT exist in y\_test
- iii) FN: the number of items in y\_test that marked as "NTFY" and do NOT exist in X\_test
- iv) TN the number of items in y\_test that do NOT marked as "NTFY" and do NOT exist in X\_test

```
precision = TP / (TP + FP)
recall = TP / (TP + FN)
```

#### b) Results

- i) We have approximately 150 items in train dataset and 20 items in test dataset
- ii) Entities metrics:

```
NTFY
precision 0.9482758620689655
recall 0.9016393442622951
TIME
precision 0.967741935483871
recall 0.9523809523809523
DATE
precision 0.9879518072289156
recall 0.91111111111111
```

#### c) Progress steps

i) Problems:

NTFY: sometimes takes dates in account, do not treat "please", do not contain Geographical items

DATE: in a week is treated as "a week ago", does not understand "next week", works badly with explicit dates like "5th of..."

TIME: in 30 minutes treats as "30 min ago"

ii) Updated dataset, added new items, enlarged the number of items and the number of epochs (from 20 to 30). Problems are solved.

## 5) Working examples

- a) Remind me today at six thirty to visit my friends
- b) I plan to visit France tomorrow at 7am
- c) Feed my cat the next week
- d) Remind me to send a message on the April 5th, 2024 at 12:25
- e) Please, check the bank account tomorrow at 3pm
- f) Remind me to walk with my dog in 30 minutes

# 6) Notes

a) be careful while testing on your own sentences. They must have dots (.) in the end.